

# THE ORBITAL PERIOD PRIOR FOR SINGLE TRANSITS

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The finite baseline of any photometric survey means that transiting planets (or stars) can have a sufficiently long period to impart just a single transit in our observations. With multiple consecutive transits, the determination of the object’s orbital period,  $P$ , is trivial - given by the time difference between events. In such cases, the photometric data,  $\mathcal{D}$ , provides a highly precise and unambiguous period, which means that choice of period prior,  $\Pr(P)$ , has negligible influence on the results. With a single transit, however, the period is only weakly constrained by the data, via the transit duration, stellar density and (unknown) eccentricity (Kipping 2010). Accordingly, the choice of period prior has a large influence on the *a-posteriori* period.

Whilst there are only a few dozen examples of single transits observed by *Kepler* amongst the 4000+ planetary candidates (Wang et al. 2015; Uehara et al. 2016; Foreman-Mackey et al. 2016), *TESS* is expected to find over a thousand (Villanueva et al. 2018) amongst a similar sized total (Barclay et al. 2018), bringing greater attention to this issue.

The prior on  $P$  should reflect the expected distribution of orbital periods as seen by the survey in question. This does not equal the intrinsic distribution of periods,  $\Pr(P)$ , because the survey’s own observational effects sculpt the distribution. For example, it would be impossible to ever record a single transit in a continuous photometric time series of baseline  $W$  with  $P < W/2$  (else multiple transits would occur), even though the intrinsic prior  $\Pr(P)$  can have support here. Thus, the fact that only a single transit occurred represents a piece of information which we can condition our prior upon, which we represent by the symbol  $\hat{n}_1$ .

In addition to  $\hat{n}_1$ , we also know the observational window,  $W$ , and the fact the planet has the correct geometry to undergo transits,  $\hat{b}$ , which biases us towards short periods more likely to display transits. Finally, we know the mid-transit time minus the start of the observational baseline,  $L$ , which we might expect to provide a useful lower limit on the period. Ignoring detection bias, which is only relevant for low signal-to-noise transits (Kipping & Sanford 2016), the overall prior on a single transit period may be expressed as

$$\Pr(P|\hat{n}_1, W, \hat{b}, L) \propto \Pr(L|\hat{n}_1, W, \hat{b}, P) \Pr(P|\hat{n}_1, W, \hat{b}) \quad (1)$$

where the right-hand side expands the left using Bayes’ theorem. The final term can be expanded in a similar way as

$$\Pr(P|\hat{n}_1, W, \hat{b}) \propto \Pr(\hat{n}_1|P, W, \hat{b}) \Pr(P|W, \hat{b}) \quad (2)$$

where  $\Pr(P|W, \hat{b})$  has no dependency on  $W$  and thus becomes  $\Pr(P|\hat{b})$ . Expanding  $\Pr(P|\hat{b})$  once more with Bayes’ theorem and then combining all the terms we have

$$\underbrace{\Pr(P|\hat{n}_1, W, \hat{b}, L)}_{\text{overall prior}} \propto \underbrace{\Pr(L|\hat{n}_1, W, \hat{b}, P)}_{\text{phase constraint}} \underbrace{\Pr(\hat{n}_1|P, W, \hat{b})}_{\text{window effect}} \underbrace{\Pr(\hat{b}|P)}_{\text{geometric bias}} \underbrace{\Pr(P)}_{\text{intrinsic prior}} \quad (3)$$

Let us write the intrinsic prior as  $P^\alpha$ , where  $\alpha = -1$  returns a log-uniform distribution. The geometric bias is well-understood and scales as  $P^{-2/3}$  after applying Kepler’s Third Law to Equation 2 of [Kipping \(2014\)](#). The window bias is more subtle and most easily calculated by defining it as

$$\Pr(\hat{n}_1|P, W, \hat{b}) = \Pr(n \geq 1|P, W, \hat{b})(1 - \Pr(n \geq 2|P, W, \hat{b})), \quad (4)$$

where  $n$  is the number of transits occurring in the window  $W$ . The first term is easy to compute by considering that  $P < W$  is guaranteed to provide at least one transit and  $P > W$  will drop off thereafter as  $1/P$ :

$$\Pr(n \geq 1|P, W, \hat{b}) \propto \begin{cases} 1 & \text{if } P \leq W, \\ \frac{W}{P} & \text{if } P > W. \end{cases} \quad (5)$$

Similarly,  $n \geq 2$  is guaranteed if  $P \leq W/2$  but impossible if  $P > W$ . Between these two extremes one should expect a  $P^{-1}$  scaling as before, and so normalizing this intermediate regime to connect to the cases into a continuous function yields:

$$\Pr(n \geq 2|P, W, \hat{b}) = \begin{cases} 1 & \text{if } P \leq \frac{W}{2}, \\ \frac{W-P}{P} & \text{if } \frac{W}{2} < P \leq W, \\ 0 & \text{if } P > W. \end{cases} \quad (6)$$

Finally, turning to the phase constraint, for  $P > W$  we expect that  $L$  is uniformly distributed between 0 and  $W$  - no phase is more likely than any other. But, if  $P$  was slightly larger than  $W/2$ , then  $L$  could only take on a narrow range around  $W/2$  in order for the transit to avoid multiple transits. This can be encoded by writing that

$$\Pr(L|P, W, \hat{b}) = \begin{cases} \frac{\mathbb{H}[L+P-W] - \mathbb{H}[L-P]}{2P-W} & \text{if } \frac{W}{2} < P \leq W, \\ \frac{\mathbb{H}[L] - \mathbb{H}[L-W]}{W} & \text{if } P > W, \end{cases} \quad (7)$$

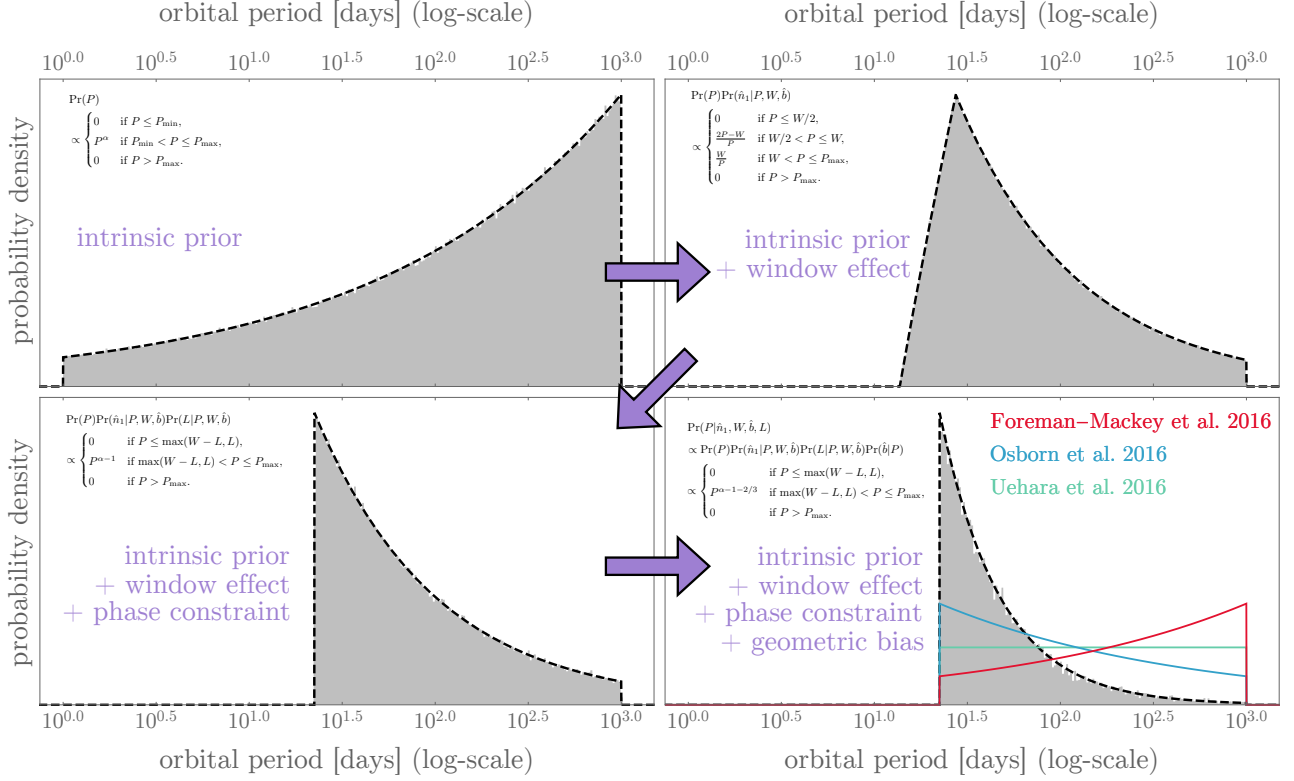
where  $\mathbb{H}$  is the Heaviside Theta function. Combining all of these components leads to the simplified result that





$$\Pr(P|\hat{n}_1, W, \hat{b}, L) \propto \begin{cases} 0 & \text{if } P \leq \max(W - L, L), \\ P^{\alpha-5/3} & \text{if } \max(W - L, L) < P \leq P_{\max}, \\ 0 & \text{if } P > P_{\max}. \end{cases} \quad (8)$$

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**Figure 1.** Comparison of  $10^6$  Monte Carlo periods (gray histograms) and the analytic formulae presented here (black dashed). Starting from the intrinsic prior (top-left), where we adopt  $\alpha = -2/3$  (log-uniform in semi-major axis) as an example, we add each component contributing to the single transit period prior one by one, assuming  $W = 27.4$  days. Initial periods are drawn using inverse transform sampling from the intrinsic distribution  and then pushed through a set of filters representing each component . All histograms are plotted in log-scale and the formulae are transformed accordingly . The bottom-left panel uses  $L = 5$  days in the Monte Carlo experiments. We compare our final distribution to three previously adopted priors for single transits in the literature ( $\sim P^{-2/3}$  by Foreman-Mackey et al. 2016;  $\sim P^{-5/3}$   by Osborn et al. 2016;  $\sim P^{-1}$  by Uehara et al. 2016).